

**An estimate of the impact of absorbing aerosol over cloud on the MODIS retrievals of
cloud optical thickness and effective radius using two independent retrievals of liquid
water path**

Eric M. Wilcox¹, Harshvardhan², and Steven Platnick¹

¹Climate and Radiation Branch, Laboratory for Atmospheres,
NASA Goddard Space Flight Center

²Department of Earth and Atmospheric Sciences, Purdue University

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Corresponding author address: Eric Wilcox, NASA Goddard Space Flight Center, Code 613.2,
Greenbelt MD 20771. (301) 614-6409 fax: (301) 614-6307. email: eric.m.wilcox@nasa.gov.

ABSTRACT

Two independent satellite retrievals of cloud liquid water path (LWP) from the NASA Aqua satellite are used to diagnose the impact of absorbing biomass burning aerosol overlaying boundary layer marine water clouds on the Moderate Resolution Imaging Spectrometer (MODIS) retrievals of cloud optical thickness (τ) and cloud droplet effective radius (r_e). In the MODIS retrieval over oceans, cloud reflectance in the 0.86 μm and 2.13 μm bands are used to simultaneously retrieve τ and r_e . A low bias in the MODIS τ retrieval may result from reductions in the 0.86 μm reflectance, which is only very weakly absorbed by clouds, owing to absorption by aerosols in cases where biomass burning aerosols occur above water clouds. MODIS LWP , derived from the product of the retrieved τ and r_e , is compared with LWP ocean retrievals from the Advanced Microwave Scanning Radiometer - EOS (AMSR-E), determined from cloud microwave emission that is transparent to aerosols. For the coastal Atlantic southern African region investigated in this study, a systematic difference between AMSR-E and MODIS LWP retrievals is found for stratocumulus clouds over three biomass burning months in 2005 and 2006 that is consistent with above-cloud absorbing aerosols. Biomass burning aerosol is detected using the ultraviolet aerosol index from the Ozone Monitoring Instrument (OMI) on the Aura satellite. The LWP difference (AMSR-E minus MODIS) increases both with increasing τ and increasing OMI aerosol index. During the biomass burning season the mean LWP difference is 14 g m^{-2} , which is within the 15-20 g m^{-2} range of estimated uncertainties in instantaneous LWP retrievals. For samples with only low amounts of overlaying smoke ($\text{OMI } AI \leq 1$) the difference is 9.4, suggesting that the impact of smoke aerosols on the mean MODIS LWP is 5.6 g m^{-2} . Only for scenes with OMI aerosol index greater than 2 does the average LWP difference and the estimated

bias in MODIS cloud optical thickness attributable to the impact of overlaying biomass burning aerosol exceed the instantaneous uncertainty in the retrievals.

1. Introduction

Satellite measurements of visible and near-infrared reflectance are now used routinely to simultaneously retrieve cloud optical thickness (τ) and cloud drop effective radius (r_e). These data are used in climate studies, among other applications, to evaluate the radiative impact of clouds on the climate system (e.g. Han et al., 1994; Quaas et al., 2006). These retrievals, however, may be biased in cases where a layer of absorbing aerosol resides above the cloud. This situation occurs frequently over the eastern South Atlantic Ocean during austral winter when smoke from extensive biomass burning in southern Africa is transported westward over a region of persistent low stratocumulus cloud cover. If the layer of absorbing biomass burning aerosol attenuates the sunlight that is both incident on and reflected by the top of the cloud, then the radiance measured by the satellite will imply a weaker cloud reflectance than that of the true cloud. This bias in estimated reflectance can result in a low bias retrieval for cloud optical thickness, and for retrievals based on certain wavelength band combinations, a biased retrieval in effective radius as well.

Haywood et al. (2004) computed the expected bias in the cloud property retrieval based on a radiative transfer model and in-situ measurements of aerosol optical properties. They report a low bias of up to 30% in τ for retrievals that depend on 0.63 or 0.86 μm reflectance values.

Biases in r_e were found to be less than 1 μm for retrievals where the visible reflectance is paired with the 3.7 or 2.13 μm reflectance. However, use of the 1.63 μm reflectance for the retrieval of r_e can result in underestimates of r_e of several μm . Biases in the retrieval of τ were greatest for bright clouds with high values of τ . Haywood et al. (2004) and Cattani et al. (2006) both present case studies from several Moderate Resolution Imaging Spectrometer (MODIS) granules indicating a bias in the 1.63 μm r_e retrieval relative to the 2.13 μm retrieval.

Bennartz (2007) found that the cloud liquid water path (*LWP*) retrieved from the Advanced Microwave Scanning Radiometer - EOS (AMSR-E) is systematically larger than the MODIS *LWP* derived from the τ and r_e retrievals for observations offshore of southern Africa during the biomass burning season; both sensors are on the NASA Aqua satellite. Bennartz and Harshvardhan (2007) argue that the difference between the AMSR-E and MODIS is qualitatively consistent with the reduced cloud optical thickness owing to absorbing aerosol overlying the cloud as estimated by Haywood et al. (2004). Here we seek to confirm this explanation by relating the difference between AMSR-E and MODIS *LWP* directly to observations of absorbing aerosol over low clouds for a study region off the coast of Angola/Namibia.

Furthermore, model simulations reported in Johnson et al. (2004) indicate that the heating above the boundary layer attributable to solar absorption by biomass burning aerosols overlaying stratocumulus clouds can enhance the strength of the inversion capping the boundary layer, thereby inhibiting cloud-top entrainment and allowing the cloud *LWP* to increase. Unraveling the impacts of aerosols on clouds using satellite data in this region requires an understanding of the impact of smoke on cloud property retrievals.

In this study we diagnose the magnitude of the *LWP* bias in the Aqua MODIS cloud property retrieval using the simultaneous and independent microwave *LWP* retrieval from AMSR-E. The microwave emission signal from clouds is not impacted by the presence of aerosol. The magnitude of the bias is reported as a function of increasing ultraviolet aerosol index determined from the nearly simultaneous Ozone Monitoring Instrument (OMI) observation on the NASA Aura satellite. Below a certain level of aerosol index, it is found that the bias in the MODIS *LWP* is less than the estimated uncertainties and the spatial/temporal RMS variability in the retrievals.

2. Methodology

110 This study makes use of two independent remote sensing retrievals of *LWP* from the NASA Aqua satellite: MODIS solar reflectance retrievals that may be susceptible to biases in cases of absorbing aerosol overlying cloud, and the AMSR-E passive microwave retrievals using wavelengths that are transparent to aerosols. We seek to estimate the impact of above-cloud absorbing biomass burning aerosol on the MODIS retrieval by investigating any systematic
115 relative bias between the two *LWP* retrievals with increasing absorbing aerosol amount above cloud. The aerosol is detected using the OMI aerosol index (*AI*), which is observed from the Aura satellite following the Aqua satellite approximately 15 minutes behind in the same orbit.

LWP and *AI* observations are obtained for the oceanic region offshore of southern Africa bounded by 10 W to 15 E longitude and 20 S to 0 S latitude during July, August, and September
120 2005 and 2006. Aerosols emanating from the burning of the African savannah are clearly evident over this region of ocean in satellite imagery (fig. 1). The OMI *AI* detects the presence of aerosols that absorb UV radiation, including in cases where the aerosols are present over stratocumulus clouds. The mean single scatter albedo of biomass burning aerosols observed during the Southern African Regional Science Initiative (SAFARI) 2000 field campaign was
125 0.91 ± 0.04 at $0.55 \mu\text{m}$ and 0.86 at $0.87 \mu\text{m}$ based on in-situ measurements from Haywood et al. (2004).

The vertical profiles of the biomass burning aerosol layers were also observed over the South Atlantic Ocean during SAFARI 2000. These aerosols were typically observed in layers that were vertically separated from stratocumulus clouds below (Hobbs, 2002; McGill et al.,
130 2003; Haywood et al., 2003). These observations imply that direct microphysical interaction

between the aerosols and stratocumulus clouds is often inhibited by the strong temperature inversion above the cloud layer.

MODIS LWP is derived from the τ and r_e retrievals in the operational Aqua MODIS Level-2 collection 5 MYD06 product. These parameters are retrieved for pixels determined to be overcast and assumed to be homogeneous (Platnick et al. 2003). A non-absorbing visible or near-infrared spectral band provides the principle sensitivity to τ . The $0.86\mu\text{m}$ reflectance is used for the ocean retrievals evaluated in this study. The absorbing near-infrared reflectance at $2.13\mu\text{m}$ provides the principle sensitivity to r_e . In situ observations of marine stratocumulus clouds in the study area indicate that their vertical structure is adiabatic with the cloud drop radius increasing with height through the cloud (Keil and Haywood, 2003). LWP for adiabatic clouds is derived from the product of r_e and τ ($LWP \approx 5/9 r_e \tau$, see Wood and Hartmann, 2006).

The AMSR-E retrieval of LWP for low clouds over ocean relies on the emission signal of condensed water using the algorithm of Wentz (1997) and Wentz and Spencer (1998). The algorithm retrieves LWP as part of a unified scheme for retrieving surface wind speed, column water vapor, LWP , and rain rate simultaneously using 5 channels of AMSR-E ranging from 6.9 GHz to 89 GHz. The retrieval is only performed over ocean because of the challenge of constraining the surface emission over land where the surface emissivity is highly variable. We use the daily 0.25° gridded products provided by Remote Sensing Systems (available from www.remss.com/amsr/amsr_browse.html).

The AMSR-E gridded LWP is an average over the entire 0.25° area, while the MODIS LWP is reported only for pixels ($\sim 1\text{km}$ at nadir) determined to be overcast. Therefore, the values are only comparable for 0.25° grid cells that are confidently determined as overcast. We average the pixels reporting a valid MODIS LWP for each 0.25 deg. grid cell, however only grid cells

with 100% of MODIS pixels reporting a valid *LWP* are included in our analysis. Because of the
 155 screening applied to the pixel-level τ and r_e retrievals, this represents a conservative screening
 for overcast conditions. Furthermore, only grid cells with average MODIS cloud-top
 temperature greater than 273 K and zero AMSR-E rain rate are included.

The presence of absorbing aerosol layers above cloud is observed with the OMI sensor on
 the Aura satellite. The OMI aerosol index is derived from the ratio between the upwelling
 160 intensity at 331 nm and the intensity at 360 nm. The index reflects the difference in the spectral
 contrast between these frequencies and the spectral contrast derived from a radiative transfer
 model of a purely Rayleigh scattering atmosphere and a Lambertian surface (Herman et al.,
 1997; Torres et al., 1998). The aerosol index as applied to OMI is given as (Ahmad et al.
 2006):

$$\text{OMI } AI = -100 \left[\log_{10} \left(\frac{I_{331}}{I_{360}} \right)_{\text{meas}} - \log_{10} \left(\frac{I_{331}}{I_{360}} \right)_{\text{calc}} \right]$$

where I_λ is the intensity at the specified wavelength and the subscripts “meas” and “calc”
 indicate the measured and model calculated quantities respectively. In a clear Rayleigh
 scattering atmosphere with low surface reflectivity (such as the ocean or land in the UV
 wavelengths considered here), the reflectivity of UV radiation in the 330 to 380 nm range
 170 decreases with increasing wavelength. Absorbing aerosols such as biomass burning emissions
 reduce the reflectivity by absorbing radiation and by reducing molecular scattering in and below
 the aerosol layer. Furthermore, UV-absorbing aerosols reduce the spectral contrast, even leading
 to cases where reflectance increases with wavelength (Hsu et al., 1996). Thus, absorbing aerosols
 lead to positive values of the OMI *AI*. *AI* increases approximately linearly with the UV optical
 175 depth with a slope that increases with decreasing single scatter albedo (Torres et al. 1998).

Clouds enhance UV reflectivity with only a weak spectral signature, and therefore do not interfere with the detection of overlying absorbing aerosol.

OMI data are obtained from the level 2 gridded products, which provide OMI *AI* footprints (13×24 km size at nadir) arranged on the 0.25° grid. Where more than one OMI *AI* footprint is centered with a single grid cell, the *AI* values are averaged. Because of the relatively large OMI footprint, some grid cells at the edge of the OMI swath do not encompass the center-point of a single OMI footprint. These grid cells are ignored in the following analysis.

Averages of the instantaneous difference between the two *LWP* retrievals are evaluated against the magnitude of the estimated uncertainties in the two retrievals. Wentz (1997) estimates the uncertainty in instantaneous microwave *LWP* retrievals (based on special sensor microwave/imager data) to be 25 g m^{-2} . More recent examinations of cloud-cleared microwave imager data indicate average systematic uncertainties in instantaneous *LWP* retrievals from 7 g m^{-2} (Greenwald et al. 2007; based on AMSR-E) to 15 g m^{-2} (Horvath and Gentemann 2007; based on TRMM Microwave Imager data). Average estimated random error in the AMSR-E data is 13 g m^{-2} (Greenwald et al. 2007). Estimated uncertainties for τ and *LWP_{MODIS}* are reported in table 1. These are provided for each pixel in the MYD06 products, and are intended as minimum uncertainties that account for estimated calibration and model uncertainty, estimated uncertainty in the surface albedo, and estimated uncertainty in the above-cloud precipitable water amount. Other error sources may be present, such as horizontal inhomogeneity, which are not accounted for. The average estimated instantaneous uncertainty in τ is 1.4 (11% of mean τ), but increases for higher values of τ (see discussion of fig. 3 below). The average estimated instantaneous uncertainty in *LWP_{MODIS}* is 19 g m^{-2} . Allowing for the possibility that only daily uncertainties are correlated reduces the estimated *LWP_{MODIS}* uncertainty to 1.4 g m^{-2} for averages

of 180 days of data. Adopting the AMSR-E values noted above from Greenwald et al. (2007)

yields a combined uncertainty for instantaneous observations of the LWP difference of $\pm 24 \text{ g m}^{-2}$. RMS variability of the LWP difference for all samples included in this study is 18 g m^{-2} (table 1).

3. Results

Figure 2 shows the LWP difference (AMSR-E minus MODIS) as a function of τ for July, August and September 2005 and 2006. Shown in the left panel are cases with OMI AI less than or equal to 1, and in the right panel are cases with OMI AI greater than 1.5. Averages and standard deviations are shown for τ bins of 2 units width. For cases of OMI $AI \leq 1$, indicating no smoke, or only small amounts of smoke overlaying the cloud, LWP_{MODIS} and LWP_{AMSR-E} agree on average to within $\pm 10 \text{ g m}^{-2}$ in most τ bins, with standard deviations representing temporal/spatial variability of approximately $\pm 20 \text{ g m}^{-2}$. The average LWP difference for these low smoke cases is 9.4 g m^{-2} . Note that these cases include a relatively small number of samples with OMI $AI < 0$, which generally indicate the presence of a scattering aerosol.

For cases with OMI $AI > 1.5$, indicating the presence of absorbing aerosol over the clouds, the LWP difference is greater than that for the $AI \leq 1$ cases and increases with increasing τ . Haywood et al. (2004) have demonstrated with a radiative transfer model that for the $0.86/2.13 \mu\text{m}$ MODIS cloud retrieval over oceans, absorption by biomass burning aerosols reduces the $0.86 \mu\text{m}$ retrieval of τ compared to the retrieval from non-polluted scenes. Furthermore, they found that the low bias in τ attributable to the influence of absorbing aerosol increases with increasing $0.86 \mu\text{m}$ reflectance; that is, the bias increases with τ . A low bias in τ would lead to a proportional low bias in LWP_{MODIS} . The results shown in figure 2 are consistent with the Haywood et al. (2004) result.

An estimate of the LWP_{MODIS} bias attributable to the smoke aerosols indicated in figure 2 is made by subtracting the average LWP difference from each τ bin under low OMI AI conditions in fig. 2a from each LWP difference sample with OMI $AI > 1.5$. The τ bias necessary to yield the LWP difference attributable to the smoke is computed using the relationship between LWP , τ and r_e described in section 2 above and displayed in figure 3. This quantity decreases with increasing τ to about -4 for clouds of τ equal to 20 to 25. The instantaneous uncertainty in the τ retrievals for these clouds is of about the same magnitude (red dashed lines), as is the RMS variability of the estimated τ bias (black error bars).

Figure 4 shows the differences between averages of high OMI AI samples and low OMI AI samples in each τ bin for LWP_{AMSR-E} (top panel), LWP_{MODIS} (middle panel) and r_e (bottom panel). LWP and τ are positively correlated. Because τ is underestimated in the high OMI AI samples owing the effect of the smoke aerosols, these samples have a higher LWP_{AMSR-E} than low OMI AI samples. The effect of the absorbing aerosol on the τ retrieval becomes apparent for clouds with τ above about 10, and increases with τ as noted above. Only a few samples have $\tau > 30$, leading to poor sampling statistics for these τ bins. Differences of 10 g m^{-2} or less are found for averages of LWP_{MODIS} between high OMI AI and low OMI AI samples in most τ bins. No difference is expected as a consequence of the bias in τ because both the retrieved τ and the inferred LWP_{MODIS} in the middle panel of fig. 4 are similarly affected. However, a significant difference might be expected if a corresponding bias were present in the r_e retrieval for the smoke-affected samples. The bottom panel in fig. 4 confirms that differences in r_e between high and low OMI AI samples are $1 \text{ }\mu\text{m}$ or less. Haywood et al. (2004) found that absorption by biomass burning aerosols introduced low biases of not more than $1 \text{ }\mu\text{m}$ in simulated retrievals with the $0.86/2.13 \text{ }\mu\text{m}$ bands used in the MODIS retrieval over oceans.

245 For comparison, LWP differences (AMSR-E minus MODIS) are shown for OMI $AI \leq 1$
 cases over the Southeast Pacific Ocean (fig. 5). In the absence of biomass burning or other
 sources of absorbing aerosol, nearly every sample in this region from the 180 day period exhibits
 OMI AI values less than or equal to 1. In contrast to the Southeast Atlantic Ocean during the
 same season, there is a systematic LWP difference for overcast marine boundary layer clouds
 250 that increasingly favors LWP_{MODIS} for increasing τ . The difference might be attributable to
 geographic differences in cloud structure. This study applies a retrieval of LWP_{MODIS} that
 assumes an adiabatic cloud structure ($LWP \approx 5/9 r_e \tau$). The LWP retrieval in the standard MODIS
 product assumes a uniform vertical cloud structure ($LWP \approx 2/3 r_e \tau$, see Stephens, 1978). When
 computing the LWP difference using the uniform cloud profile retrieval for the Southeast
 255 Atlantic clouds, a similar bias is found that increasingly favors MODIS for increasing τ (not
 shown). While changing the retrieval algorithm applied to MODIS removed this systematic
 difference for Southeast Atlantic Ocean cases (fig. 2), it did not for Southeast Pacific Ocean
 cases. Aircraft observations indicate that the effective radius of cloud drops generally increase
 through the depth of the cloud (Keil and Haywood, 2003). This increase is linear in height above
 260 cloud base in the adiabatic cloud model. It is important to note that the retrieval applied here
 assumes the retrieved r_e is at the very top of the cloud. Weighting functions for MODIS
 retrievals of r_e and τ are reported by Platnick (2000) where it is found that the retrievals include
 reflectance contributions from a layer at the top of the cloud. As a result, biases in retrieved LWP
 assuming uniform vertical cloud structure are only 3% to 10% for modeled adiabatic clouds;
 265 smaller than expected. Additionally, entrainment drying at cloud top can reduce cloud drop sizes
 near cloud top below the size expected from the adiabatic model.

The mean difference between LWP_{AMSR-E} and LWP_{MODIS} is reported in table 1 for all overcast samples and for overcast samples within various ranges of OMI AI . The average LWP difference (AMSR-E minus MODIS) is 14 g m^{-2} . For samples with only low amounts of
 270 overlaying smoke ($OMI \text{ } AI \leq 1$), however, the difference is 9.4 g m^{-2} . This difference of 5.6 g m^{-2} , attributable to the impact of aerosols on the MODIS retrieval, is less than the uncertainty in instantaneous LWP retrievals from either instrument (15 g m^{-2} and 19 g m^{-2} for AMSR-E and MODIS respectively) or the RMS variability of the LWP difference (18 g m^{-2}). Bennartz (2007) reports differences in area averaged LWP during the biomass burning season that favor LWP_{AMSR-E}
 275 E by about 10 g m^{-2} . Bennartz and Harshvardhan (2007) attribute this difference to the smoke effect on the MODIS retrieval of τ . The results reported here indicate that part of this difference is independent of the smoke effect.

The LWP difference between the highest OMI AI samples ($AI > 3$) and the lowest OMI AI samples ($AI \leq 1$) is 24.6 g m^{-2} and indicates the maximum impact of smoke aerosol on
 280 LWP_{MODIS} . The LWP difference increases systematically with increasing OMI AI . This dependence is shown in figure 6 (left panel) and suggests a low bias in the MODIS retrieval of τ with a magnitude that increases with the aerosol amount. The LWP difference attributable to the impact of the aerosols is estimated by subtracting the mean LWP difference for all samples with $OMI \text{ } AI \leq 1$ from each instantaneous sample of the LWP difference. The estimated τ bias
 285 implied by this LWP difference attributable to the aerosol impact is computed using the relationship between LWP , τ and r_e described in section 2 and shown in fig. 6b. Only for those samples with OMI AI greater than 2 does this estimated bias in τ exceed the ± 1.4 mean instantaneous uncertainty in the τ retrieval. These samples represent less than 9% of all overcast samples. The mean LWP difference for samples with $AI=2$ is 20.5 g m^{-2} , which is comparable in

290 magnitude to both the estimated instantaneous uncertainty in the MODIS retrieval and the RMS
variability of the LWP difference. Only in a relatively small number of cases does biomass
burning aerosol from Africa result in an impact on the τ and LWP retrievals of 20% or greater for
stratocumulus clouds over the South Atlantic Ocean.

Although the bias in MODIS τ and LWP retrievals owing to biomass burning aerosols
295 only exceeds the other errors in these retrievals in a relatively few cases, the bias is detectable
and present in regionally and seasonally averaged analyses (Bennartz, 2007; Bennartz and
Harshvardhan, 2007). This study indicates that with collocated observations in ultraviolet bands,
such as provided by OMI, the samples subjected to such a bias can be identified and avoided.
Alternatively, if microwave LWP observations are also available, it is possible to retain those
300 samples and estimate the magnitude of the bias.

4. Summary

Collocated MODIS, AMSR-E, and OMI data are used to identify cases of biomass
burning aerosol overlaying boundary layer stratocumulus clouds offshore of Western Africa
during the July-September biomass burning seasons of 2005 and 2006. MODIS retrievals of τ
305 and r_e are expected to be biased in cases where absorbing aerosol above a cloud layer is
sufficient to reduce the visible/near-infrared cloud reflectance at the satellite sensor. The
existence and magnitude of this bias can be estimated using simultaneous retrievals of LWP from
AMSR-E, which observes the microwave emissions from clouds that are transparent to aerosols.
 LWP_{MODIS} derived from the product of the retrieved τ and r_e is found to agree on average with
310 LWP_{AMSR-E} to within $\pm 10 \text{ g m}^{-2}$ for averages over τ bins across most of the range of observed τ .
For Southeast Pacific Ocean cases with $OMI \leq 1$, there is a systematic difference between
 LWP_{MODIS} and LWP_{AMSR-E} for overcast samples that increasingly favors LWP_{MODIS} for increasing

τ which is not associated with the presence of biomass burning aerosol. For cases with a greater amount of smoke aerosols (OMI $AI > 1.5$), LWP_{MODIS} is biased low relative to LWP_{AMSR-E} , which
315 is consistent with the expectation that the $0.86 \mu\text{m}$ MODIS radiances are reduced by the absorption in the aerosol layer, leading to a low bias in the τ retrieval. The r_e retrieval is only minimally impacted by the aerosol. The effect of absorbing biomass burning aerosols on the LWP_{MODIS} retrieval is 5.6 g m^{-2} on average, which is within the estimated instantaneous uncertainty for the LWP retrievals and within the RMS variability of individual samples of the
320 LWP difference. The magnitude of the τ bias increases with τ and also increases with the magnitude of the OMI AI . Only for cases with OMI $AI > 2$, which account for about 9% of overcast samples, does the bias in LWP_{MODIS} relative to LWP_{AMSR-E} and the estimated bias in τ attributable to the impact of biomass burning aerosols exceed the instantaneous uncertainty of the retrievals.

5. References

- Ahmad, S. P., O. Torres, P. K. Bhartia, G. Leptoukh, and S. Kempler (2006), Aerosol Index from TOMS and OMI Measurements. *Proceedings of the 14th AMS conference on the Applications of Air Pollution Meteorology with the Air and Waste Management Assoc.*
- Bennartz, R. (2007), Global assessment of marine boundary layer cloud droplet number concentration from satellite. *J. Geophys. Res.*, 112, D02201, doi:10.1029/2006JD007547.
- Bennartz, R. and Harshvardhan (2007), Correction to “Global assessment of marine boundary layer cloud droplet number concentration from satellite”. *J. Geophys. Res.*, D16302, doi:10.1029/2007JD008841.
- Cattani, E., M. J. Costa, F. Torricella, V. Levizzani, and A. M. Silva (2006), Influence of aerosol particles from biomass burning on cloud microphysical properties and radiative forcing. *Atmos. Res.*, 82, 310-327.
- Greenwald, T. J., T. S. L’Ecuyer, and S. A. Christopher (2007), Evaluating specific error characteristics of microwave-derived cloud liquid water products. *Geophys. Res. Lett.*, 34, doi:10.1029/2007GL031180.
- Han, Q., W. B. Rossow, and A. A. Lacis (1994), Near-global survey of effective droplet radii in liquid water clouds using ISCCP data. *J. Climate*, 7, 465-497.
- Haywood, J. M., S. R. Osborne, P. N. Francis, A. Keil, P. Formenti, M. O. Andreae, and P. H. Kaye (2003), The mean physical and optical properties of regional haze dominated by biomass burning aerosol measured from the C-130 aircraft during SAFARI 2000. *J. Geophys. Res.*, 108(D13), 8473, doi:10.1029/2002JD002226.
- Haywood, J. M., S. R. Osborne, and S. J. Abel (2004), The effect of overlying absorbing aerosol layers on remote sensing retrievals of cloud effective radius and cloud optical depth. *Q. J. R. Meteorol. Soc.*, 130, 779-800.
- Herman, J. R., P. K. Bhartia, O. Torres, C. Hsu, C. Seftor, and E. Celarier (1997), Global distribution of UV-absorbing aerosols from Nimbus 7/TOMS data. *J. Geophys. Res.*, 102(D14), 16 911-16 922.
- Hobbs, P. V. (2002), Clean air slots amid atmospheric pollution. *Nature*, 415, 861.
- Horváth, Á and R. Davies (2007), Comparison of microwave and optical cloud water path estimates from TMI, MODIS, and MISR. *J. Geophys. Res.*, 112, D01202, doi:10.1029/2006JD07101.
- Horváth, Á. and C. Gentemann (2007), Cloud fraction dependent bias in satellite liquid water path retrievals of shallow non-precipitating marine clouds. *Geophys. Res. Lett.*, 34, doi:10.1029/2007GL030625.
- Hsu, N. C., J. R. Herman, P. K. Bhartia, C. J. Seftor, O. Torres, A. M. Thompson, J. F. Gleason, T. F. Eck, and B. N. Holben (1996) Detection of biomass burning smoke from TOMS measurements. *Geophys. Res. Lett.*, 23, 745-748.
- Johnson, B. T., K. P. Shine, P. M. Forster, The semi-direct aerosol effect: Impact of absorbing aerosols on marine stratocumulus, *Q. J. R. Meteorol. Soc.*, **130**, 1407-1422, 2004.

- 1 Keil, A. and J. M. Haywood (2003), Solar radiative forcing by biomass burning aerosol particles
2 during SAFARI 2000: A case study based on measured aerosol and cloud properties. *J.*
3 *Geophys. Res.*, 108, doi:10.1029/2002JD002315.
- 4 McGill, M. J., D. L. Hlavka, W. D. Hart, E. J. Welton, and J. R. Campbell (2003), Airborne lidar
5 measurements of aerosol optical properties during SAFARI-2000. *J. Geophys. Res.*, 108,
6 doi:10.1029/2002JD002370.
- 7 Platnick, S. (2000), Vertical photon transport in cloud remote sensing problems. *J. Geophys.*
8 *Res.*, 105, 22 919-22 935.
- 9 Platnick, S., M. D. King, A. Ackerman, W. P. Menzel, B. A. Baum, J. C. Riédi, and R. A. Frey
10 (2003), The MODIS cloud products: Algorithms and examples from Terra. *IEEE Trans.*
11 *Geosci. Rem. Sens.*, 41, 459-473.
12
- 13 Quaas, J., O. Boucher, and U. Lohmann (2006), Constraining the total aerosol indirect effect in
14 the LMDZ and ECHAM4 GCMs using MODIS satellite data. *Atmos. Chem. Phys.*, 6,
15 947-955.
- 16 Stephens, G. L. (1978), Radiation profiles in extended water clouds. II: Parameterization
17 schemes. *J. Atmos. Sci.*, 35, 2123-2132.
- 18 Torres, O., P. K. Bhartia, J. R. Herman, Z. Ahmad, and J. Gleason (1998) Derivation of aerosol
19 properties from satellite measurements of backscattered ultraviolet radiation: Theoretical
20 basis. *J. Geophys. Res.*, 103(D14), 17 099-17 110.
- 21 Wentz, F. J. (1997), A well-calibrated ocean algorithm for special sensor microwave / imager. *J.*
22 *Geophys. Res.*, 102, 8703-8718.
- 23 Wentz, F. J. and R. W. Spencer (1998), SSM/I rain retrievals within a unified all-weather ocean
24 algorithm. *J. Atmos. Sci.*, 55, 1613-1627.
- 25 Wood, R. and D. L. Hartmann (2006), Spatial variability of liquid water path in marine low
26 cloud: The importance of mesoscale cellular convection, *J. Climate*, **19**, 1748-1764.
- 27

Table 1. LWP_{AMSR-E} , LWP difference ($LWP_{AMSR-E} - LWP_{MODIS}$), τ uncertainty, and LWP_{MODIS} uncertainty statistics for all overcast samples and overcast samples of various OMI AI ranges. LWP_{MODIS} uncertainty assumes all pixel-level uncertainties are correlated. LWP_{MODIS} uncertainty values in parentheses assume only daily uncertainties are correlated while uncertainties from different days are uncorrelated. RMS values of LWP difference are enclosed in parentheses. All units are g m^{-2} except for number of samples and τ uncertainty.

	all AI	$AI \leq 1$	$1 < AI \leq 2$	$2 < AI \leq 3$	$AI > 3$
number of samples	43,140	21,206	17,012	4354	568
mean LWP_{AMSR-E}	92	90	92	95	103
median LWP_{AMSR-E}	80	80	80	80	100
est. τ uncertainty	1.4	1.4	1.4	1.4	1.2
est. LWP_{MODIS} uncertainty	19 (1.4)	20 (1.5)	18 (1.4)	17 (1.3)	16 (1.2)
mean LWP difference (RMS)	14 (18)	9.4 (15)	16 (19)	23 (23)	34 (25)

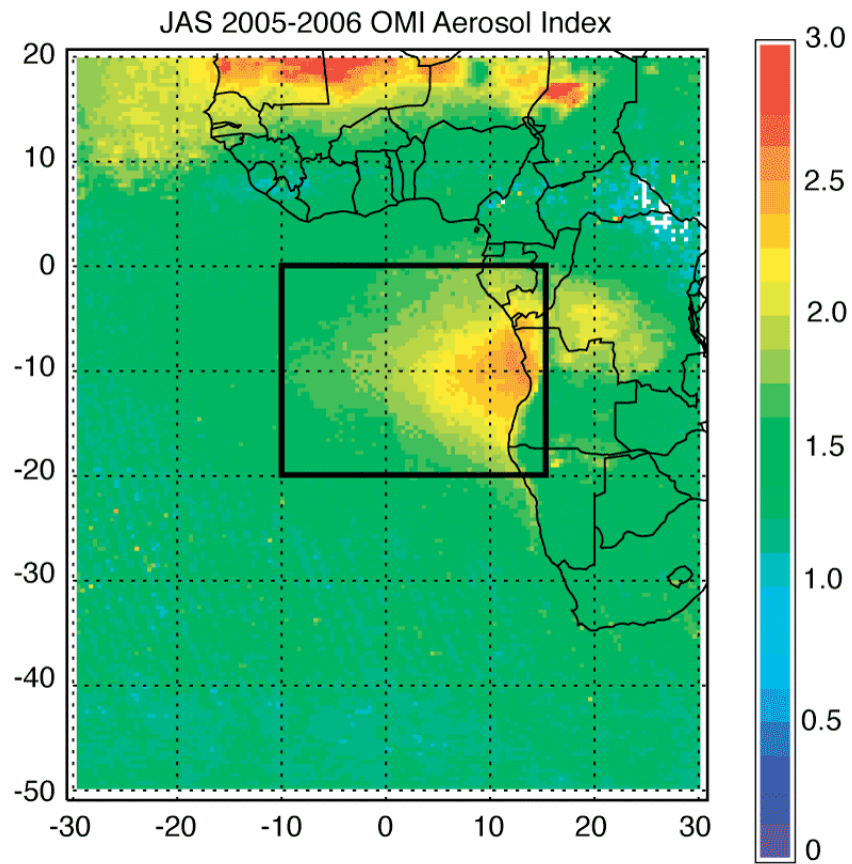


Fig. 1. 2005-2006 July, August, and September average OMI aerosol index. The box indicates the geographic bounds applied to the data in this analysis.

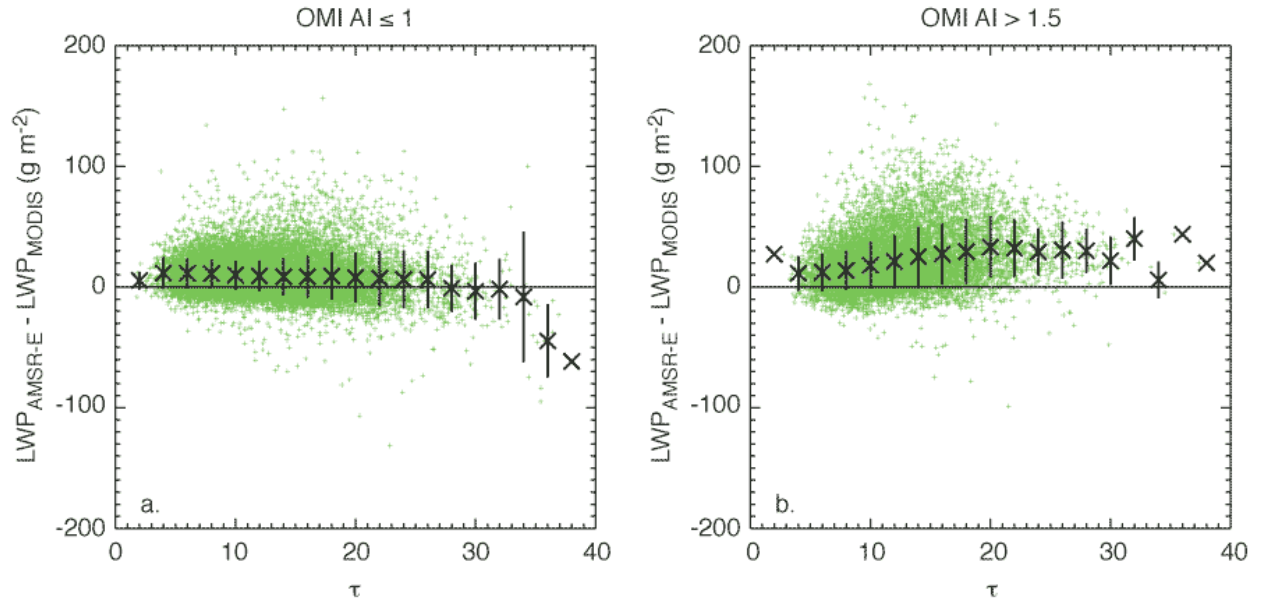


Fig. 2. *LWP* difference (AMSRE – MODIS) against MODIS cloud optical thickness for all overcast samples. Green pluses are individual 0.25 deg. gridded samples. X-symbols are averages for τ bins of 2 units width. Vertical lines indicate standard deviations of τ bin averages. For each pair, left panel is samples corresponding to OMI $AI \leq 1.0$ and right panel is OMI $AI > 1.5$.

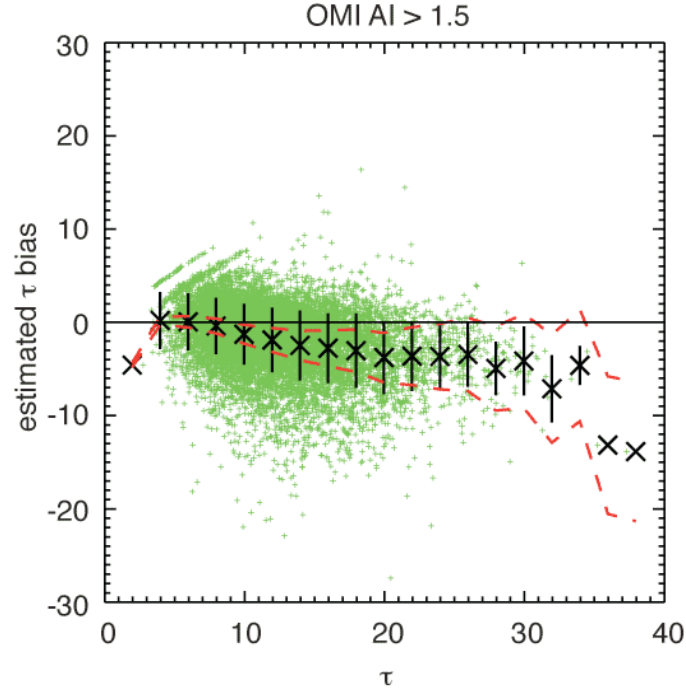


Fig. 3. Estimated bias in MODIS cloud optical thickness for samples with OMI AI > 1.5 against τ for all overcast samples. Green pluses are individual 0.25 deg. gridded samples. X-symbols are averages for τ bins of 2 units width. Vertical lines indicate standard deviations of τ bin averages. Red dashed lines indicate range of estimated instantaneous uncertainty in the MODIS τ retrieval.

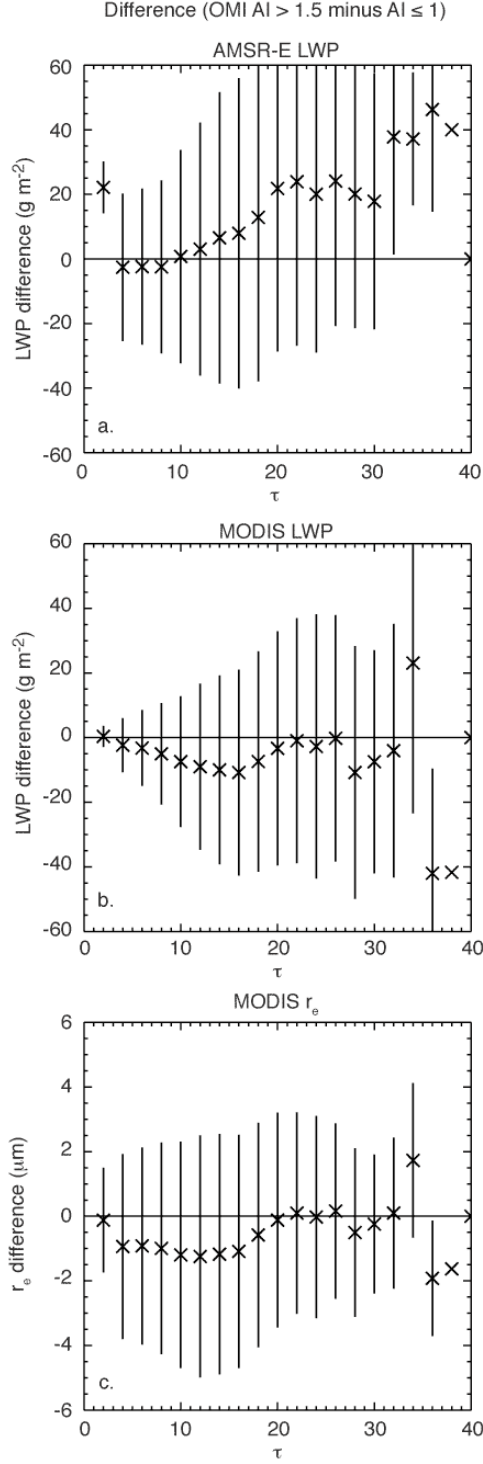
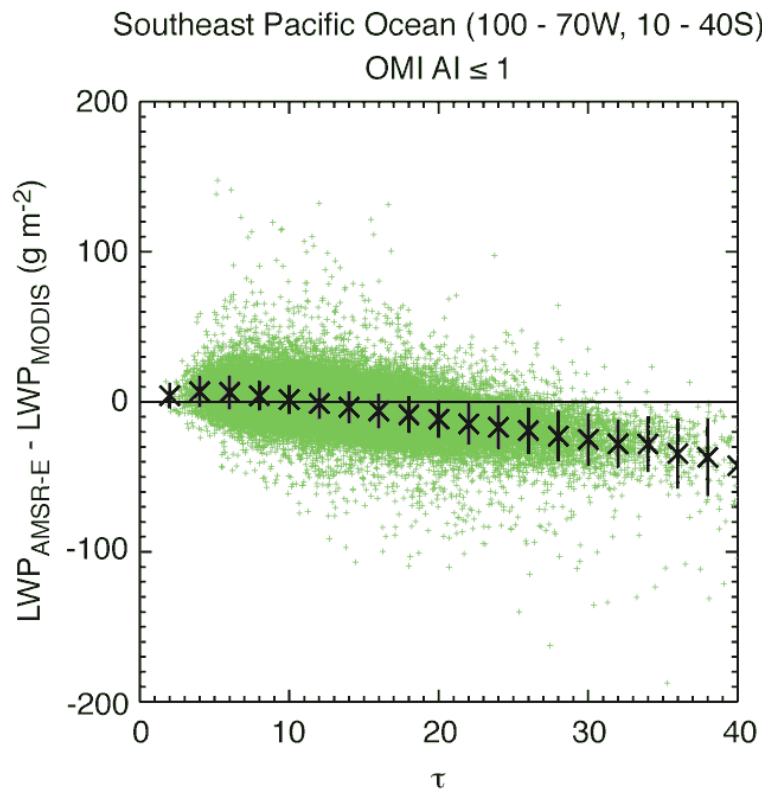


Fig.4. LWP and r_e differences between high OMI AI samples ($AI > 1.5$) and low OMI AI samples ($AI \leq 1$) against MODIS cloud optical thickness. X-symbols are differences of averages over τ bins of 2 units width. Vertical lines are the sum (in quadrature) of the standard deviation of the clean cases and the polluted cases in each τ bin. Top panel is LWP_{AMSR-E} , middle panel is LWP_{MODIS} , and bottom panel is r_e (MODIS).

1



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4 Fig.5. 2005-2006 July, August, and September LWP difference (AMSR-E – MODIS) against τ
 5 for all overcast samples in South Pacific stratocumulus region with OMI AI ≤ 1 (only a
 6 few samples exhibited OMI AI values > 1). Green pluses are individual 0.25 deg. gridded
 7 samples. X-symbols are averages τ bins of 2 units width. Vertical lines indicate standard
 8 deviations of τ bin averages.

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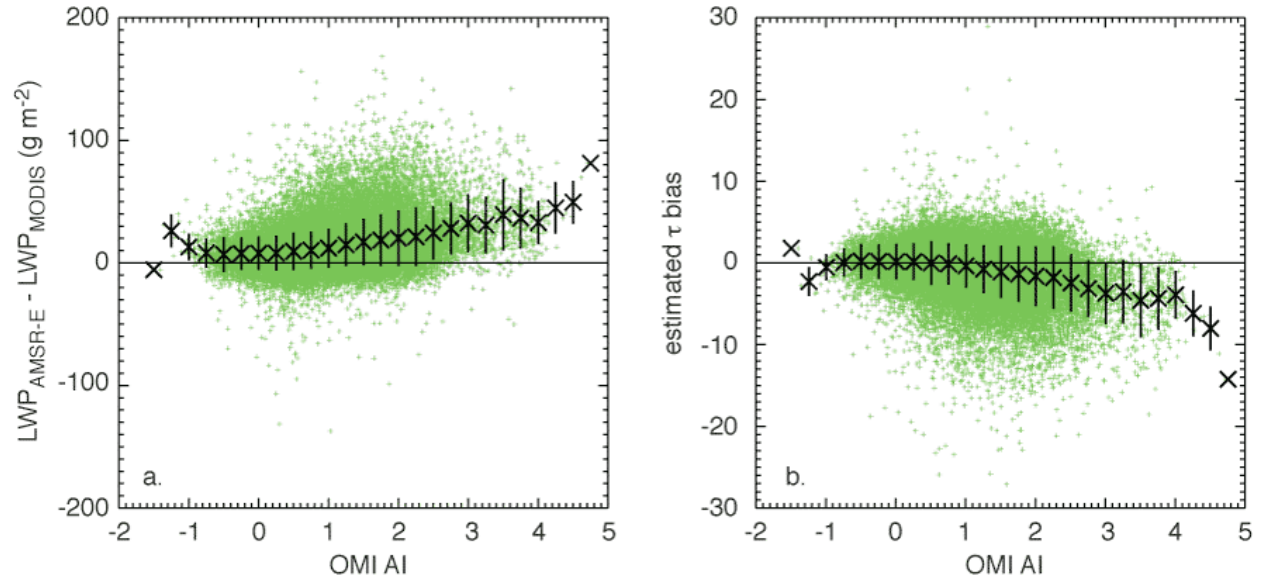


Fig. 6. LWP difference (AMSRE – MODIS; left panel) and estimated τ bias (right panel) against OMI AI for all overcast samples. Green pluses are individual 0.25 deg. gridded samples. X-symbols are averages for OMI AI bins of 0.25 magnitude width. Vertical lines indicate standard deviations of OMI AI bin averages.